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1 %% Machine Learning
2 % Lab 10: Spam Classification with SVMs
3 % — Water overflow —
4 %{
5 In the first half of the exercise, you will implement regularized linear
6 regression to predict the amount of water flowing out of a dam using the
7 change of water level in a reservoir. In the next half, you will go through
8 some diagnostics of debugging learning algorithms and examine the effects
9 of bias v.s. variance.
10 %}
11
12 %% Initialization
13 clear ; close all; clc
14
15 %% ===== Part 1: Email Preprocessing =====
16 % To use an SVM to classify emails into Spam v.s. Non-Spam, you first need
17 % to convert each email into a vector of features. In this part, you will
18 % implement the preprocessing steps for each email. You should
19 % complete the code in processEmail.m to produce a word indices vector
20 % for a given email.
21
22 fprintf('\nPreprocessing sample email (emailSample1.txt)\n');
23
24 % Extract Features
25 file_contents = readFile('emailSample1.txt');
26 word_indices = processEmail(file_contents);
27
28 % Print Stats
29 fprintf('Word Indices: \n');
30 fprintf(' %d', word_indices);
31 fprintf('\n\n');
32
33 fprintf('Program paused. Press enter to continue.\n');
34 pause;
35
36 %% ===== Part 2: Feature Extraction =====
37 % Now, you will convert each email into a vector of features in R^n.
38 % You should complete the code in emailFeatures.m to produce a feature
39 % vector for a given email.
40
41 fprintf('\nExtracting features from sample email (emailSample1.txt)\n');
42
43 % Extract Features
44 file_contents = readFile('emailSample1.txt');
45 word_indices = processEmail(file_contents);
46 features = emailFeatures(word_indices);
47
48 % Print Stats
49 fprintf('Length of feature vector: %d\n', length(features));
50 fprintf('Number of non-zero entries: %d\n', sum(features > 0));
51
52 fprintf('Program paused. Press enter to continue.\n');
53 pause;
54
55 %% ===== Part 3: Train Linear SVM for Spam Classification =====
56 % In this section, you will train a linear classifier to determine if an

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57 % email is Spam or Not-Spam.
58
59 % Load the Spam Email dataset
60 % You will have X, y in your environment
61 load('spamTrain.mat');
62
63 fprintf('\nTraining Linear SVM (Spam Classification)\n')
64 fprintf('(this may take 1 to 2 minutes) ...\n')
65
66 C = 0.1;
67 model = svmTrain(X, y, C, @linearKernel);
68
69 p = svmPredict(model, X);
70
71 fprintf('Training Accuracy: %f\n', mean(double(p == y)) * 100);
72
73 %% ===== Part 4: Test Spam Classification =====
74 % After training the classifier, we can evaluate it on a test set. We have
75 % included a test set in spamTest.mat
76
77 % Load the test dataset
78 % You will have Xtest, ytest in your environment
79 load('spamTest.mat');
80
81 fprintf('\nEvaluating the trained Linear SVM on a test set ...\n')
82
83 p = svmPredict(model, Xtest);
84
85 fprintf('Test Accuracy: %f\n', mean(double(p == ytest)) * 100);
86 pause;
87
88
89 %% ===== Part 5: Top Predictors of Spam =====
90 % Since the model we are training is a linear SVM, we can inspect the
91 % weights learned by the model to understand better how it is determining
92 % whether an email is spam or not. The following code finds the words with
93 % the highest weights in the classifier. Informally, the classifier
94 % 'thinks' that these words are the most likely indicators of spam.
95 %
96
97 % Sort the weights and obtain the vocabulary list
98 [weight, idx] = sort(model.w, 'descend');
99 vocabList = getVocabList();
100
101 fprintf('\nTop predictors of spam: \n');
102 for i = 1:15
103     fprintf(' %-15s (%f) \n', vocabList{idx(i)}, weight(i));
104 end
105
106 fprintf('\n\n');
107 fprintf('\nProgram paused. Press enter to continue.\n');
108 pause;
109
110 %% ===== Part 6: Try Your Own Emails =====
111 % Now that you've trained the spam classifier, you can use it on your own
112 % emails! In the starter code, we have included spamSample1.txt,
113 % spamSample2.txt, emailSample1.txt and emailSample2.txt as examples.

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114 % The following code reads in one of these emails and then uses your
115 % learned SVM classifier to determine whether the email is Spam or
116 % Not Spam
117
118 % Set the file to be read in (change this to spamSample2.txt,
119 % emailSample1.txt or emailSample2.txt to see different predictions on
120 % different emails types). Try your own emails as well!
121 filename = 'spamSample1.txt';
122
123 % Read and predict
124 file_contents = readFile(filename);
125 word_indices = processEmail(file_contents);
126 x = emailFeatures(word_indices);
127 p = svmPredict(model, x);
128
129 fprintf('\nProcessed %s\n\nSpam Classification: %d\n', filename, p);
130 fprintf('(1 indicates spam, 0 indicates not spam)\n\n');

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gaussianKernel.m

```

1 function sim = gaussianKernel(x1, x2, sigma)
2 %RBFKERNEL returns a radial basis function kernel between x1 and x2
3 %   sim = gaussianKernel(x1, x2) returns a gaussian kernel between x1 and x2
4 %   and returns the value in sim
5
6 % Ensure that x1 and x2 are column vectors
7 x1 = x1(:); x2 = x2(:);
8
9 % You need to return the following variables correctly.
10 sim = 0;
11
12 sim = exp(-(sum((x1-x2).^2)/(2*sigma^2)));
13
14 end

```

dataset3Params.m

```

1 function [C, sigma] = dataset3Params(X, y, Xval, yval)
2 %DATASET3PARAMS returns your choice of C and sigma for Part 3 of the exercise
3 %where you select the optimal (C, sigma) learning parameters to use for SVM
4 %with RBF kernel
5
6 % You need to return the following variables correctly.
7 C = 1;
8 sigma = 0.3;
9
10 maxError = Inf;
11
12 % It would be nice if this didn't do function optimization by exhaustive search :-/
13 for currC = [0.01 0.03 0.1 0.3 1 3 10 30]
14     for currSigma = [0.01 0.03 0.1 0.3 1 3 10 30]
15         model = svmTrain(X, y, currC, @(x1, x2) gaussianKernel(x1, x2, currSigma));
16
17         predictions = svmPredict(model, Xval);
18         predictionError = mean(double(predictions ~= yval));
19
20         if predictionError < maxError
21             maxError = predictionError;
22             C = currC;
23             sigma = currSigma;
24         end
25     end
26 end
27
28 end

```

processEmail.m

```

1 function word_indices = processEmail(email_contents)
2 %Preprocesses the body of an email and returns a list of indices of the
3 %words contained in the email.
4
5 % Load Vocabulary
6 vocabList = getVocabList();
7
8 % Init return value
9 word_indices = [];

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10
11 % ===== Preprocess Email =====
12
13 % Find the Headers ( \n\n and remove )
14 % Uncomment the following lines if you are working with raw emails with the
15 % full headers
16
17 % hdrstart = strfind(email_contents, ([char(10) char(10)]));
18 % email_contents = email_contents(hdrstart(1):end);
19
20 % Lower case
21 email_contents = lower(email_contents);
22
23 % Strip all HTML
24 % Looks for any expression that starts with < and ends with > and replace
25 % and does not have any < or > in the tag it with a space
26 email_contents = regexprep(email_contents, '<[^\>]+\>', ' ');
27
28 % Handle Numbers
29 % Look for one or more characters between 0-9
30 email_contents = regexprep(email_contents, '[0-9]+', 'number');
31
32 % Handle URLs
33 % Look for strings starting with http:// or https://
34 email_contents = regexprep(email_contents, ...
35                             '(http|https)://[^\s]*', 'httpaddr');
36
37 % Handle Email Addresses
38 % Look for strings with @ in the middle
39 email_contents = regexprep(email_contents, '[^\s]+@[^\s]+', 'emailaddr');
40
41 % Handle $ sign
42 email_contents = regexprep(email_contents, '[$]+', 'dollar');
43
44
45 % ===== Tokenize Email =====
46
47 % Output the email to screen as well
48 fprintf('\n==== Processed Email ==== \n\n');
49
50 % Process file
51 l = 0;
52
53 while ~isempty(email_contents)
54
55     % Tokenize and also get rid of any punctuation
56     [str, email_contents] = ...
57         strtok(email_contents, ...
58             [' @$/#.-:&*+=[ ]?!(){},">_<;%' char(10) char(13)]);
59
60     % Remove any non alphanumeric characters
61     str = regexprep(str, '[^a-zA-Z0-9]', '');
62
63     % Stem the word
64     % (the porterStemmer sometimes has issues, so we use a try catch block)
65     try str = porterStemmer(strtrim(str));
66     catch str = ''; continue;

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67     end;
68
69     % Skip the word if it is too short
70     if length(str) < 1
71         continue;
72     end
73
74     % Look up the word in the dictionary and add to word_indices if
75     % found
76     % =====
77     for i = 1:length(vocabList)
78         if(strcmp(str, vocabList{i}))
79             word_indices = [word_indices ; i];
80         end
81     end
82     % =====
83
84     % Print to screen, ensuring that the output lines are not too long
85     if (l + length(str) + 1) > 78
86         fprintf('\n');
87         l = 0;
88     end
89     fprintf('%s ', str);
90     l = l + length(str) + 1;
91
92 end
93
94 % Print footer
95 fprintf('\n\n=====\\n');
96
97 end

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emailFeatures.m

```

1  function x = emailFeatures(word_indices)
2  %Takes in a word_indices vector and
3  %produces a feature vector from the word indices.
4
5  % Total number of words in the dictionary
6  n = 1899;
7
8  % You need to return the following variables correctly.
9  x = zeros(n, 1);
10
11  for i = 1:length(word_indices)
12      x(word_indices(i))=1;
13  end
14
15  end

```